

Subtotal ordering – a pedagogically advantageous algorithm for computing total degree reverse lexicographic order

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Abstract

Total degree reverse lexicographic order is currently generally regarded as most often fastest for computing Gröbner bases. This article describes an alternate less mysterious algorithm for computing this order using exponent subtotals and describes why it should be very nearly the same speed the traditional algorithm, all other things being equal. However, experimental evidence suggests that subtotal order is actually slightly faster for the *Mathematica*[®] Gröbner basis implementation more often than not. This is probably because the weight vectors associated with the natural subtotal weight matrix and with the usual total degree reverse lexicographic weight matrix are different, and *Mathematica* also uses those the corresponding weight vectors to help select successive S polynomials and divisor polynomials: Those selection heuristics appear to work slightly better more often with subtotal weight vectors.

However, the most important advantage of exponent subtotals is pedagogical. It is easier to understand than the total degree reverse lexicographic algorithm, and it is more evident why the resulting order is often the fastest known order for computing Gröbner bases.

Keywords: Term order, Total degree reverse lexicographic, tdeg, grevlex, Gröbner basis

1 Introduction

Total degree reverse lexicographic order (degRevLex) is currently generally regarded as most often fastest for computing Gröbner bases..¹ This order is usually determined by Algorithm 1. However, as indicated in the comments therein, this algorithm tends to be mystifying and therefore difficult to recall:

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¹The adjective “graded” is sometimes used instead of “total degree”.

Algorithm 1 degRevLex order of two exponent vectors

Given: Nonnegative integer exponent vectors $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_n]$, $\beta = [\beta_1, \beta_2, \dots, \beta_n]$.
Returns: One of “ \prec ”, “ $=$ ”, or “ \succ ” according to the degRevLex order between power products $z_1^{\alpha_1} z_2^{\alpha_2} \dots z_n^{\alpha_n}$ and $z_1^{\beta_1} z_2^{\beta_2} \dots z_n^{\beta_n}$ with indeterminate order $z_1 \succ z_2 \succ \dots \succ z_n$.

```
a ← α1;  
b ← β1;  
for k ← 2 to n do  
    a ← a + αk;  
    b ← b + βk; end for;  
if a < b, then return “ $\prec$ ”;  
if a > b, then return “ $\succ$ ”; /* : So far this all makes sense. */  
for k ← n to 1 by −1 do  
    if αk > βk, then return “ $\prec$ ”; /* : Whoa! Why is “ $\succ$ ” matched with “ $\prec$ ”?  
    if αk < βk, then return “ $\succ$ ”; end for; /* : These must be typographical errors!”  
return “ $=$ ”; /* : At least this step makes sense! */
```

This is the order that Buchberger in his Ph.D. dissertation and the order that Gröbner always used when discussing multivariate polynomials (Buchberger, personal communication).²

The next fastest widely discussed order is total degree lexicographic order, and it too orders primarily by total degree. Therefore clearly total degree is very important for speed, and that makes sense because if we are iteratively annihilating terms with the largest total degrees, then the degree of each variable can’t increase beyond that total degree.

Consequently, it seems plausible that it would be more consistent to break total-degree ties with the sum of the degrees of all but the least main variable, then break those ties with the sum of the degrees of all but the two least main variables, and so on, as described in Algorithm 2.

²Trinks [8] made the extraordinarily useful contribution of introducing the idea of admissible orderings and the particularly useful alternative example of lexicographic ordering.

Algorithm 2 subtotal order of two exponent vectors

Given: Nonnegative integer exponent vectors $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_n]$, $\beta = [\beta_1, \beta_2, \dots, \beta_n]$.

Returns: One of “ \prec ”, “ $=$ ”, or “ \succ ” according to the subtotal order between power products $z_1^{\alpha_1} z_2^{\alpha_2} \dots z_n^{\alpha_n}$ and $z_1^{\beta_1} z_2^{\beta_2} \dots z_n^{\beta_n}$ with indeterminate order $z_1 \succ z_2 \succ \dots \succ z_n$.

```
A1 ← α1;  
B1 ← β1;  
for k ← 2 to n do  
    Ak ← Ak-1 + αk;  
    Bk ← Bk-1 + βk; end for;  
for k ← n to 1 by -1 do  
    if Ak > βk then return “ $\succ$ ”; /* : I have a better feeling about this! */  
    if Ak < βk then return “ $\prec$ ”; end for;  
return “ $=$ ”;
```

Algorithms 1 and 2 both do $2n$ additions followed by up to n comparisons, with the same looping costs.

Both algorithms assume that the variables have been extracted from the power products, which requires that all power products have the same number of exponents even if some of these exponents are 0. If instead the power products have only non-zero exponents, and therefore also contain variables to indicate the base for each exponent, then the algorithms are slightly different but both of them still have the same complexity as each other.

Section 2 discusses weight matrices, weight vectors, and their implications for subtotal *versus* degRevLex order. Section 3 describes the experimental procedures for comparing these two orders and the results of those comparisons, with conclusions in Section 4.

2 Weight matrices and weight vectors

When I first thought of subtotal order, I wondered if subtotal order would be faster than degRevLex order. Therefore I wanted a quick way to compare them experimentally without having to implement my own Gröbner basis algorithm or learn the intricacies of an existing one to modify it. Fortunately some computer algebra systems permit users to specify their own ordering merely by providing a non-singular real square *weight matrix* W : To compare exponent vectors α and β for ordering, we lexicographically compare corresponding *weight vectors*

$$\begin{aligned} \mathbf{w}(\alpha) &\leftarrow [w_1(\alpha), w_2(\alpha), \dots] \leftarrow W \cdot \alpha^T, \\ \mathbf{w}(\beta) &\leftarrow [w_1(\beta), w_2(\beta), \dots] \leftarrow W \cdot \beta^T. \end{aligned}$$

The weight matrix corresponding to algorithm 2 is

$$W_{\text{sub}} = \begin{pmatrix} 1 & 1 & \dots & 1 & 1 \\ 1 & 1 & \dots & 1 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & \dots & 0 & 0 \\ 1 & 0 & \dots & 0 & 0 \end{pmatrix}. \quad (1)$$

Proposition 1. *The ordering specified by W_{sub} is an admissible ordering.*

Proof. The first element of each column in W_{sub} is positive.

Also, matrix W_{sub} is non-singular because the columns, hence the variables, can be permuted into an upper triangular matrix having 1 for every diagonal element, and the determinant of a square upper triangular matrix is the product of its diagonal elements, which consequently is non-zero, making the permuted W_{sub} , hence also W_{sub} non-singular. These are sufficient conditions for an admissible ordering. \square

Every admissible term order can be associated with a weight matrix, and the one that is usually associated with algorithm 1 for degRevLex ordering is

$$W_{\text{degRevLex}} = \begin{pmatrix} 1 & 1 & \dots & 1 & 1 \\ 0 & 0 & \dots & 0 & -1 \\ 0 & 0 & \dots & -1 & 0 \\ \vdots & \vdots & & \vdots & \vdots \\ 0 & -1 & \dots & 0 & 0 \end{pmatrix}. \quad (2)$$

There are at least two ways to use weight matrices in an implementation:

1. Do a matrix-vector multiplication for a power product every time we want to order it relative to another power product. To save some time, we could interleave the matrix-vector multiplication with the comparison of successive weight vector components to quit as soon as a difference is detected.
2. Do matrix-vector multiplications only for the given *input* polynomials, store the weight vectors in parallel with the corresponding exponent vectors, then
 - (a) whenever two power products are multiplied, compute the weight vector of their product as the sum of the two weight vectors, and
 - (b) whenever two power products are divided, compute the weight vector of their quotient as the difference of the two weight vectors.

At the expense of some additional storage space, the second choice is clearly much faster, because the number of power product comparisons during computation of a Gröbner basis is usually far more than the number of power products in the given polynomials.

Mathematica accepts weight matrices, and it uses the second of these two methods – even for the implemented built-in orderings, for which it uses essentially Algorithm 1

rather than $W_{\text{degRevLex}}$ to initialize the weight vectors in the case of degRevLex ordering. Thus we can expect built-in degRevLex order for that implementation to be slightly faster than providing $W_{\text{degRevLex}}$, but usually not dramatically so.

Nonetheless, to make the comparison fair – more as if comparing built-in degRevLex ordering to built-in subtotal ordering with initialization via Algorithm 2 – I started timing a few examples done with both W_{sub} and $W_{\text{degRevLex}}$. I soon noticed that both matrices always gave the same Gröbner basis. This could have been a coincidence. However:

Proposition 2. *Subtotal ordering is equivalent to degRevLex.*

Proof. The first rows of W_{sub} and $W_{\text{degRevLex}}$ are identical. The second row of W_{sub} given by formula (1) is the second row of $W_{\text{degRevLex}}$ given by formula (2) plus the first row of $W_{\text{degRevLex}}$. The third row of W_{sub} is the first row of $W_{\text{degRevLex}}$ plus the second and third rows of $W_{\text{degRevLex}}$, and so on. Thus $W_{\text{degRevLex}}$ can be transformed into W_{sub} by a sequence of adding positive multiples of rows to rows below them. This is a sufficient condition for orderings specified by two weight matrices to be equivalent. \square

Remark. After noticing the identical Gröbner bases I discovered that Chee-Keng Yap [10] describes the computation of degRevLex ordering by Algorithm 1, but lists W_{sub} rather than the usual $W_{\text{degRevLex}}$ as a weight matrix for degRevLex ordering. He didn't explain why, but I suspect that he had the idea of subtotals too.

With weight matrices $W_{\text{degRevLex}}$ and W_{sub} inducing the same ordering, we should expect their speed ratios to be close to 1.0.

Section 3 indicates that this is often so, but not always, then explains why there are exceptions in the case of *Mathematica*.

3 Experimental procedures and results

The discussion in Sections 1 and 2 suggests that subtotal order should be very nearly the same speed as degRevLex order – regardless of whether they are both determined directly from exponent vectors or both determined via weight matrices and perhaps also weight vectors. This section describes experimental procedures and results that test this hypothesis.

3.1 Experimental procedures

A *Mathematica* function invocation of the form

```
TimeConstrained[Timing[
  GroebnerBasis[{polynomial1, polynomial2, ...}, {variable1, variable2, ...},
  Sort → True, MonomialOrder → DegreeReverseLexicographic];],
  maximumSeconds]
```

computes a degRevLex Gröbner basis of the polynomials after heuristically reordering the variables for speed, then displays only the computing time – or displays **\$Aborted** if *maximumSeconds* is exceeded.

The rewrite rule

```
SubtotalWeightMatrix[n_] := Table[Table[If[i ≤ n - j + 1, 1, 0], j, n], i, n]
```

defines a function that returns an n by n weight matrix for subtotal ordering.

The rewrite rule

```
DegRevLexWeightMatrix[n_] :=  
Table[Table[If[i == 1, 1, If[i == n + 2 - j, -1, 0]], {j, n}], {i, n}]
```

defines a function that returns an n by n weight matrix for subtotal ordering.

Therefore a *Mathematica* function invocation of the form

```
TimeConstrained[Timing[  
GroebnerBasis[{polynomial1, polynomial2, ...}, {variable1, ..., variablen},  
Sort → True, MonomialOrder → SubtotalWeightMatrix[n]];],  
maximumSeconds]
```

computes a subtotal Gröbner basis of the polynomials after heuristically reordering the variables for speed, then displays only the computing time or `$Aborted`. A similar function invocation using `DegRevLexWeightMatrix` displays the computation time of a `degRevLex` basis using a weight matrix.

I didn't want to address the issue of inexact computation at this time, because it is handled quite differently by different systems.³ Therefore I avoided test cases that contain Floats, unless they were obviously representations of rational numbers having small magnitude denominators, in which case I rationalized those Floats.

I wanted to detect any difference in the relative speeds of term-order comparisons, and coefficient arithmetic tends to be a larger portion of the total computing time when Gröbner bases are computed over the rational numbers rather than over the integers modulo a prime whose square fits in one computer word. Therefore, I did all problems in the coefficient domain \mathbb{Z}_{32003} , because the prime 32003 maps only about 0.003% of all non-zero integer coefficients to 0, but 32003 is small enough so that its square fits within one 32-bit computer word.

Some of the original test cases and the ones obtained by rationalizing simple floating-point coefficients contained rational coefficients that weren't integers. Such polynomials were multiplied by the least common multiple of their coefficient denominators so that computing a Gröbner basis in \mathbb{Z}_{32003} was straightforward.

My patience for typing examples, checking for typographical errors, and waiting for results is limited. Also, I wanted to avoid the extra space of listing previously unpublished examples in two-dimensional .pdf format, forcing others to do lengthy error-prone typing to try all of these examples on some other Gröbner basis implementation. Therefore I searched the Internet for medium-sized examples – preferably available in text that I could copy, paste and quickly edit to replace semicolons with commas, etc; and I used 120

³Many implementations make no special effort for Floats, making the results disastrously sensitive to differences in floating point arithmetic and differences in the order of operations.

seconds as the time limit. Many such examples are available at [9]. Others are available at [6] and [1]. These sites also give original references for the examples. This article lists input polynomials for a few additional examples that I think have not previously been published.

Although the `Sort → True` parameter should make the results rather insensitive to the order in $\{variable_1, \dots, variable_n\}$, I also entered the list of variables in the order specified in the sources or that I could infer, in case it mattered. Some of the original problems involved eliminating some of the variables or treating some as parameters in the coefficient domain. However, for uniformity I simply computed the Gröbner basis with respect to all of the variables.

My objective was to compare the speed of subtotal versus degRevLex ordering algorithms. To make the comparison fair, I used a weight matrix for degRevLex too. However, I also used the built-in degRevLex option to estimate how much improvement to expect if the subtotal ordering algorithm was built in.

The computer has a 1.60GHz Intel Core 2 Duo U9600 CPU with 3 gigabytes of RAM.

The Windows Vista operating system appears to have a timer resolution of only about 0.015 seconds. Therefore if a time was less than 1 second, then I issued the command

```
Timing [Do [
  GroebnerBasis [{polynomial1, polynomial1, ...}, {variable1, variable1, ...},
  Sort → True, MonomialOrder → ...], {m}]];
```

with the number of repetitions m sufficient to make the time exceed 1 second, then divided by m to obtain the time for one repetition. Despite this, even with no network connection and no voluntary programs launched other than *Mathematica*, times tend to vary upon repetition within an interval of about $\pm 3\%$ of their mean. *Mathematica* uses reference counts rather than garbage collection, so this variation is probably caused instead by the multiple core architecture and the irregular competing activity of the many resident programs that lurk in most Windows installations.

Relative speed is most important for problems that require extensive time. Therefore of all the problems that I tried, I included the ones that took the most time for built-in degRevLex order without exceeding the time limit for any ordering algorithm – as many examples as fit in a one-page table. The examples that I thus excluded didn't have noticeably different overall behavior regarding the relative speed of subtotal *versus* degRevLex ordering algorithms.

3.2 The test examples

Here are the included examples that I believe aren't already publicly published in some form:

1. Lichtblau 1:⁴

$$\begin{aligned} & t^4 z b + x^3 y a, \\ & t x^8 y z - a b^4 c d e, \\ & x y^2 z^2 d + z c^2 e^2, \\ & t x^2 y^3 z^4 + a b^2 c^3 e^2, \end{aligned}$$

with variable order $\{t, x, y, z, a, b, c, d, e\}$.

2. Lichtblau 2:

$$\begin{aligned} & a^2 + b^2 + 2c^2 + 2d^2 + 3f^2 + 3g^2 - h, \\ & 70ab + 140ac + 140bd + 28cd + 252cf + 252dg + 18fg - 105u, \\ & 28bc + 28c^2 + 28ad + 28d^2 + 42af + 12df + 24f^2 + 42bg + 12cg + 24g^2 - 35v, \\ & 36cd + 30bf + 24cf + 30ag + 24dg + 16fg - 35x, \\ & 8df + 2f^2 + 8cg + 2g^2 - 7y, \\ & 100fg - 77z, \end{aligned}$$

with variable order $\{a, b, c, d, f, g, h, u, v, x, y, z\}$.

3. Lichtblau 3:

$$\begin{aligned} & -3375uv + 3291u^2v + 750uv^2 - 732u^2v^2 + 2225uv^3 - 2209u^2v^3 + Qx, \\ & P + Qy, \\ & P + Qz, \\ & r(1 + 24uv + 24u^2v - 24uv^2), \end{aligned}$$

where

$$\begin{aligned} P &= -400u + 350u^2 - 4800uv + 4800u^2v + 7425uv^2 - 7359u^2v^2 - 2225uv^3 + 2209u^2v^3, \\ Q &= (1 + 24uv + 24u^2v - 24uv^2), \end{aligned}$$

with variable order $\{x, y, z, u, v, r\}$.

4. A geometry problem of Michael Trott:

$$\begin{aligned} & -x_1 + x_2 + y_1 - 3x_1^2y_1 + 2x_1^3y_1 - 2x_1y_1^3 + 2x_2y_1^3 - y_2 + 3x_1^2y_2 - 2x_1^3y_2, \\ & x_1 - x_2 - y_1 + 3x_2^2y_1 - 2x_2^3y_1 + y_2 - 3x_2^2y_2 + 2x_2^3y_2 + 2x_1y_2^3 - 2x_2y_2^3, \\ & -1 + 2x_1 - 2x_1^3 + x_1^4 + 2y_1 + y_1^4, \\ & -1 + 2x_2 - 2x_2^3 + x_2^4 + 2y_2 + y_2^4, \\ & 1 + (x_1 - x_2)z + (y_1 - y_2)z^2, \end{aligned}$$

with variable order $\{x_1, x_2, y_1, y_2, z\}$.

5. *Mathematica* Help - GröbnerBasis - Options - Sort:

$$\begin{aligned} & 3x^7 + 5xyz^2 - 10y^2z - 6xz + y^3 + w, \\ & -2x^2z + 3x^3y^2 + y^4 - 12xz - 8xz^2 + 3y^2z - 11wxy^2, \\ & 10x^2w - 7yzw^2 - 2xz^4w + 4x^2y + 3xy^2 - 6yz^3 - w + 2, \\ & w^3 - wx^2y + xyz^2 - 2wxz^2 - 3w - 2xy^2 - 3, \end{aligned}$$

with variable order $\{w, x, y, z\}$.

⁴The problems provided by Daniel Lichtblau are from a mix of literature, user questions and bug reports, with unrecorded individual provenance.

6. Variation on a theme of Giovini *et. al*: One of my test files is so similar to Giovini 3.7 [4] that I must have entered it twice, but once with typographical errors. These seemingly minor changes make the example that is relatively nearly fastest for subtotal ordering algorithm become an example that is relatively slowest for that algorithm. The variation is

$$\begin{aligned} &x^{33}z^{23} - y^{82}a, \\ &x^{45} - y^{13}z^{21}b, \\ &x^{41}c - y^{33}z^{12}, \\ &x^{22} - y^{33}z^{12}d, \\ &x^5y^{17}z^{22}e - 1, \\ &xyzt - 1, \end{aligned}$$

with variable order $\{t, b, c, e, d, a, z, x, y\}$.

All of the other examples are already published elsewhere, as cited in Table 1.

3.3 Test results

The last column of Table 1 displays the most important results – the ratios of the computing time for subtotal ordering versus degRevLex ordering, both using a weight matrix.

Notice that the rows are ordered by non-decreasing values in this last column in an attempt to discern correlations with the number of variables and/or total degrees of the input polynomials. These total degrees are also listed in the Table with, for example, $6^3 \cdot 5$ meaning 3 polynomials each having total degree 6 and 1 polynomial having total degree 5. However, the number of examples and/or their variability doesn't appear to be large enough to reveal obvious correlations. This might be partly because the standard deviation of the number of variables is only 2.8 for a median and mean of about 8.

Most of the speed ratios are rather close to 1.0, as expected. However, there are some notable outliers at the top and bottom of the table. A probable explanation for these outliers is that in *Mathematica* the weight vectors are also used to select the next S-polynomial or reducer polynomial. If so, then it is an indication that good selection of a next S-polynomial and a next reducer polynomial should instead if possible be based on the *ordering* that the weight vector induces, so as to be invariant to a property that doesn't correlate completely with ordering.

Although re-executing an example caused time variations for individual examples within an interval of about $\pm 3\%$, there are enough examples so that the summary statistics are more tightly repeatable.

The median speed ratio of 0.98 and mean of 0.92 suggest that the subtotal algorithm would be about this much faster than the degRevLex algorithm if both were built into *Mathematica*. The standard deviation of 0.24 weakens that conclusion, but most of the standard deviation comes from the outliers at the top of the table where the subtotal algorithm was significantly faster.

The speeds of the two methods are close enough so that quite possibly the degRevLex algorithm could be faster than the subtotal algorithm if both were built into another Gröbner basis implementation that used weight vectors in a different way to select S

polynomials and reducers. However, it seems highly likely that the speeds would be quite close if weight vectors weren't used for selection strategy.

The penultimate column of Table 1 is next most interesting:

- The summary statistics for that column weakly support the conclusion that built-in degRevLex is faster than the degRevLex weight matrix by only 2% for the median or 8% for the mean. This can only be attributable to computing the initial weight vectors from the weight matrices, and these small percentages indicate that the initialization of weight vectors is usually only a small portion of the total time.
- However, there are a surprising number of instances where the weight matrix is slightly *faster* – up to 7%. I can think of no implementation reason why this should be. If there is no such reason, then it can be taken as an indication of the repeatability deviations in *individual* time ratios – up to 7% rather than the 3% that I estimated from informal experiments.

4 Conclusions

The main result is that subtotal ordering is an alternate way to view degRevLex ordering that more clearly explains its good behavior.

A secondary conclusion is that the algorithm for computing subtotal order is very nearly the same speed as that for computing degRevLex order – at least if the induced ordering rather than the weight vectors is used to select the next S-polynomial and next reducer.

The relative speeds of the subtotal and degRevLex ordering algorithms are close enough so that it probably isn't worth replacing the degRevLex algorithm with the subtotal algorithm in existing implementations. However, it is well worth considering use of the subtotal algorithm instead of the degRevLex algorithm in new implementations.

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Table 1: Coefficient in \mathbb{Z}_{32003} . Seconds & time ratios for subtotal *vs* degRevLex order

search term & citations	# vars	input tot. degrees	grevlex builtin sec.	$\frac{\text{grevlex}}{(\text{grevlex})_{\text{matrix}}}$	$\frac{\text{subtotal}}{(\text{grevlex})_{\text{matrix}}}$
Cohn3 [1]	4	$6^3 \cdot 5$	7.24	0.08	0.08
filter design [6]	9	$5, 4^2 \cdot 3, 2^4$	10.1	0.98	0.22
benchmark_i1 [1]	10	3^{10}	2.78	0.30	0.31
Assur44 [1]	8	$3^3 \cdot 2^5$	6.30	0.40	0.38
Giovini 3.7 [4]	9	$83 \cdot 45^3 \cdot 44 \cdot 4$	36.7	0.95	0.65
benchmark_D1 [1]	12	$3^2 \cdot 2^9 \cdot 1$	0.71	0.59	0.73
des22_24 [1, 9]	10	$2^8 \cdot 1^2$	0.75	0.70	0.73
Lichtblau 2	9	$11 \cdot 10 \cdot 6^2$	0.44	0.72	0.90
Gonnet <i>et. al.</i> [5]	17	2^{19}	6.01	0.74	0.95
cdpm5 [1]	5	3^5	4.60	0.95	0.95
reimer5 [9]	5	$6 \cdot 5 \cdot 4 \cdot 3 \cdot 2$	1.59	0.99	0.96
Kotsireas4body [6]	6	$5^3 \cdot 2^3$	2.92	1.05	0.97
Lichtblau 3	12	2^6	0.50	0.88	0.97
cyclic6 [1, 9]	6	$6 \cdot 5 \cdot 4 \cdot 3 \cdot 2, 1$	0.62	1.01	0.97
Giovini 3.1 [4]	7	$4^2 \cdot 3^{10} \cdot 2$	0.57	1.17	0.97
eco8 [1]	8	$3^3 \cdot 2 \cdot 1$	1.18	0.95	0.98
redeco7 [9]	8	$2^6 \cdot 1^2$	0.67	0.96	0.98
extcyc5 [1]	6	$5^2 \cdot 4 \cdot 3 \cdot 2 \cdot 1$	1.15	1.00	0.98
f744 [1]	12	$3^2 \cdot 2^2 \cdot 1^2$	5.13	0.93	0.98
Lichtblau 1	6	5^3	5.91	1.02	0.99
virasoro [9]	8	2^8	23.8	0.99	1.00
Trott geometry	5	$4^3 \cdot 3$	12.4	0.99	1.00
Kotsireaus4bodySymmetric [6]	7	$5^2 \cdot 3^2 \cdot 2^2$	36.2	1.00	1.01
Katsura7 [9]	7	$2^6 \cdot 1$	0.78	0.99	1.01
redcyc6 [9]	6	$11 \cdot 5 \cdot 4 \cdot 3 \cdot 2 \cdot 1$	0.43	1.00	1.01
Harrier RK2 [?]	13	$4^4 \cdot 3^2 \cdot 2 \cdot 1^4$	33.5	0.95	1.03
<i>Mathematica</i> help	4	$7 \cdot 6 \cdot 5 \cdot 4$	5.98	1.01	1.04
rpb124 [9]	9	$3^2 \cdot 2^6 \cdot 1$	1.87	1.07	1.04
Tran, rational implicitization [7]	5	$6^2 \cdot 5$	5.13	0.96	1.05
kinema [1]	9	$2^6 \cdot 1^3$	2.07	1.03	1.07
Kotsireaus5body [6, 9]	6	$5^3 \cdot 2^3$	3.18	1.03	1.10
rpbl [9]	6	$3^5 \cdot 2$	0.94	1.07	1.13
variation on Giovini 3.7	9	$83 \cdot 46 \cdot 45^3 \cdot 4$	19.8	0.94	1.39
Statistics↓					
Median	8.0		2.85	0.98	0.98
Mean	8.1		7.34	0.92	0.92
Standard Deviation	2.8		10.7	0.19	0.24
Count < 1.0000				22	20
Count > 1.0000				10	11